

# Amortized analysis

# Amortized analysis

- In an amortized analysis, we estimate the time required to perform a sequence of operations.
- With amortized analysis, we have the opportunity to show that the average cost of an operation is small, even if the cost of such operation looks big (according to asymptotic notation)

# Amortized analysis

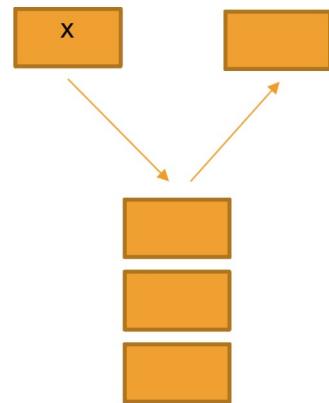
Two techniques

- Accounting method
  - Faster to use, intuitive. Better suited for simple operations
- Potential method
  - More formal and structured. Better suited for involved algorithms

# First example

Consider a stack data structure

- Push ( $S, x$ ) –  $O(c) \approx \delta(1)$ 
  - Input:
    - $S$  stack *(= constant)*
    - $x$  element to insert on top of  $S$
- Pop ( $S$ ) –  $O(c) \approx \delta(1)$ 
  - Input:
    - $S$  stack
  - Output
    - $x$  element removed from the top of  $S$



Stack: Last in, first out

## First example

Suppose you want to run n operations of type POP and PUSH.

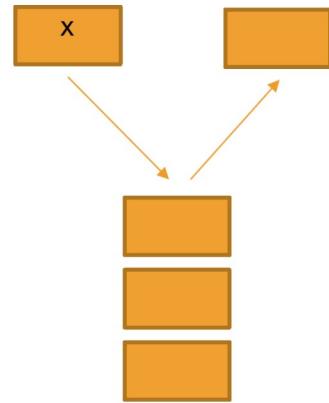
What will be the total cost?

Pick the maximum cost among the operations you may run.

$O(c)$

Multiply for the times you are running these operations

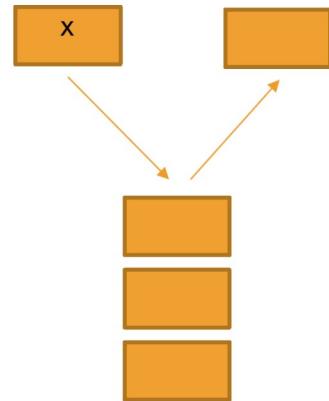
$O(n)$



# First example

Suppose we add a new operation to our stack

```
function MULTIPOP (S, k){  
    input  
    - S stack data structure  
    - k, number of elements to remove from S  
    while (not S.empty() and k>0){  
        POP(S)  
        k = k-1  
    }  
}
```



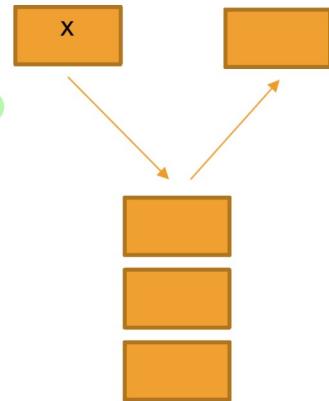
What is the cost of the new operation?

$O(k)$   
K - parameter integer, not constant

## First example

Suppose you want to run  $n$  operations of type POP, PUSH and MULTIPOP.

What will be the total cost?



Pick the maximum cost among the operations you may run.

$$O(k)$$

Multiply for the times you are running these operations

$$O(kn) = O(n^2)$$

# Accounting Method

- In the **accounting method** of amortized analysis, we assign differing charges to different operations, with some operations charged more or less than they actually cost. We call the amount we charge an operation its **amortized cost**.
  - We assign the difference to specific objects in the data structure as **credit**.
  - We must ensure that the total amortized cost of a sequence of operations provides an upper bound on the total actual cost of the sequence.

# Accounting method

	REAL COST	SCHEMA	
POP	1	0	$O(c)$
PUSH	1	2	$O(c)$
MULTIPOP	$k$	0	$O(c)$

(number of pops)

$$\sum_{i=1}^n \hat{c}_i \geq \sum_{i=1}^n c_i$$

schema

Real cost

push gets 2 credits because  
pop & multipop rely on something bc  
pushed

Pick the maximum cost among the operations you  
may run.

$O(c)$

Multiply for the times you are running these operations  
 $O(n)$

Schema works as worst-case for all  
operations.

## Dynamic table

- We implement a dynamic array that can be resized every time there is no space for a new element



## Dynamic table - algorithm

- Simple insert –  $O(c)$ 
  - When the array has an empty spot
- Complex insert –  $O(k)$  when no empty  
  - Resize + insert need to copy K times
  - When the array is full

How does it take to run n inserts?

$$O(k)*n = O(n)*n = O(n^2)$$

# Accounting method

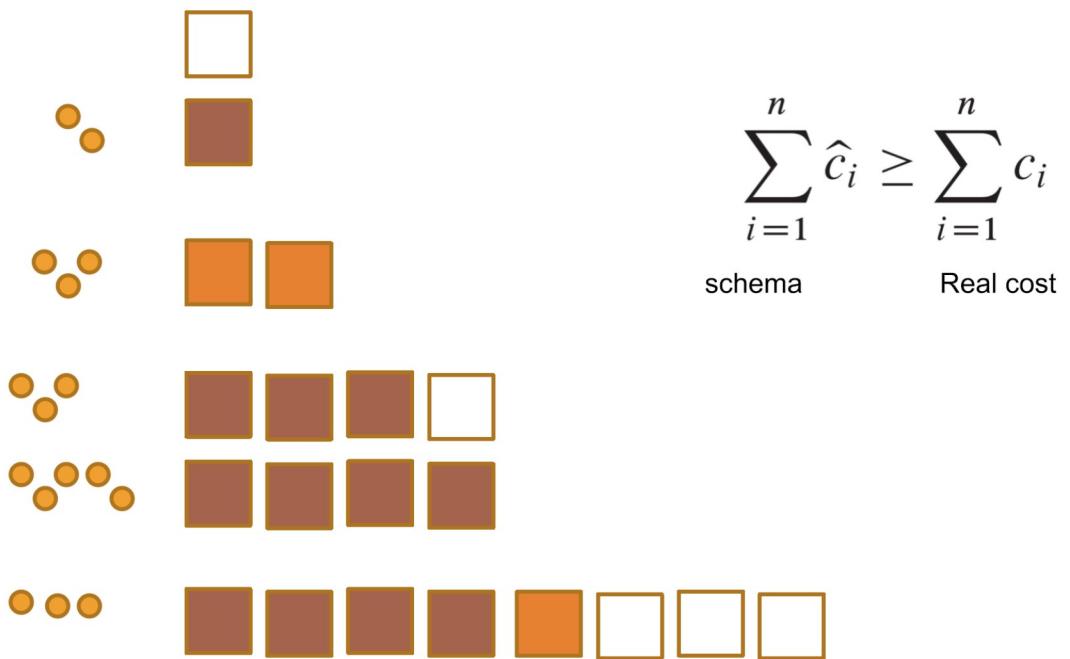
	Real cost operation	Schema	
Insert	1	3	
Resize	k	0	

Pick the maximum cost among the operations you may run.

O(c)

Multiply for the times you are running these operations  
O(n)

# Accounting method



## What's next?

- Book chapters
  - Chapter 1.4
- In class activities
  - Q&A session
- Next class
  - Amortized analysis with the potential method

